ON STEADY

The Bright Side of Mathematics



Probability Theory - Part 24

for

tion: Let
$$(X_{t})_{t\in T}$$
 be a stochastic process with $T\subseteq \mathbb{Z}$ or $T\subseteq \mathbb{R}$.
We call $(X_{t})_{t\in T}$ Markov process or Markov chain if
for all $n \in \mathbb{N}$, $t_{1}, t_{2}, ..., t_{n}, t \in T$, $t_{1} < t_{2} < \cdots < t_{n} < t$,
and $x_{1}, x_{2}, ..., x_{n}, x \in \mathbb{R}$, we have:
 $\mathbb{P}(X_{t} = x \mid X_{t_{1}} = x_{1}, X_{t_{2}} = x_{2}, ..., X_{t_{n}} = x_{n})$
 $= \mathbb{P}(X_{t} = x \mid X_{t_{n}} = x_{n})$
discrete-time Markov chain:
 $X_{n} \land X_{t_{n}} = x_{n}$
 $depends only on X_{n}, x, t_{n}
 $f_{x,y}(k, k+1) = \mathbb{P}(X_{k+1} = y \mid X_{k} = x)$$

$$\rho_{x,y}(k,k+1) = \left| P(X_{k+1} = y \mid X_k = x) \right|$$

transition probability
from x to y at time k
time = k
time = k+1

If $\rho_{x,y}(k,k+1)$ does not depend on k, then we say:

the Markov chain is time-homogeneous



Here:
$$f = \begin{pmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 0 & 1 \end{pmatrix}$$

Start the game with $q^0 = (1, 0, 0) \xrightarrow{\text{one time-step}} q^1 = (\frac{1}{2}, \frac{1}{2}, 0)$

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one time-step
$$q^{2} = \left(\frac{1}{2}, \frac{1}{4}, \frac{1}{4}\right)$$

$$q^2 = q^1 P$$
 (vector-matrix-multiplication)
 $\Rightarrow q^n = q^0 P^n$ Law of total probability
 $q^{n \to \infty}$
 $(0, 0, 1)$?